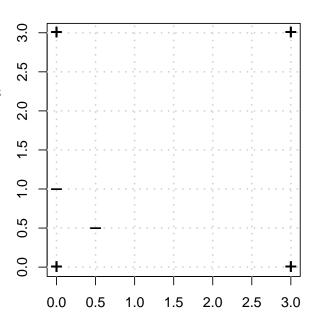
## Exercise 1: Soft Margin Classifier

The primal optimization problem for the two-class soft margin SVM classification is given by

$$\begin{aligned} \min_{\theta, \theta_0, \zeta^{(i)}} \quad & \frac{1}{2} ||\theta||^2 + C \sum_{i=1}^n \zeta^{(i)} \\ \text{s.t.} : \quad & y^{(i)} (\theta^\top \mathbf{x}^{(i)} + \theta_0) \ge 1 - \zeta^{(i)}, \\ & \zeta^{(i)} \ge 0, \quad \forall i = 1, \dots, n. \end{aligned}$$



- (a) Add the decision boundary to the figure for  $\hat{\theta} = (1,1)^T$ ,  $\hat{\theta}_0 = -2$ . (NB: This is the approximate optimum for C = 10)
- (b) Identify the coordinates of the points on the margin hyperplanes and compute the values of their slack variables  $\zeta^{(i)}$ .

## Exercise 2: Optimization

Write your own stochastic subgradient descent routine to solve the soft-margin SVM in the primal formulation.

Hints:

- Use the regularized-empirical-risk-minimization formulation, i.e., an optimization criterion without constraints
- No kernels, just a linear SVM.
- Compare your implementation with an existing implementation (e.g., kernlab in R). Are your results similar? Note that you might have to switch off the automatic data scaling in the already existing implementation.